

Predictability in Space Launch Vehicle Anomaly Detection Using Intelligent Neuro-Fuzzy Systems

JPL

JSC

McDonnell Douglas
Joint Effort

Lockheed

JPL Team

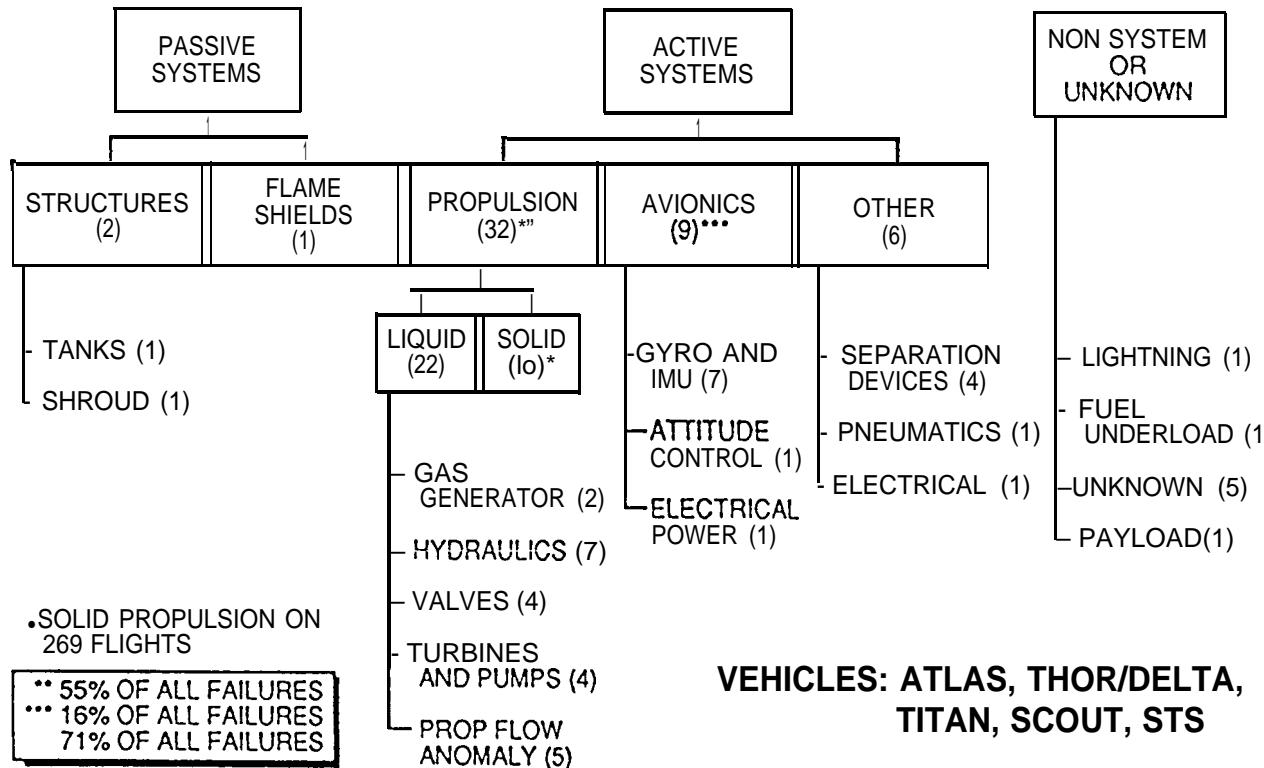
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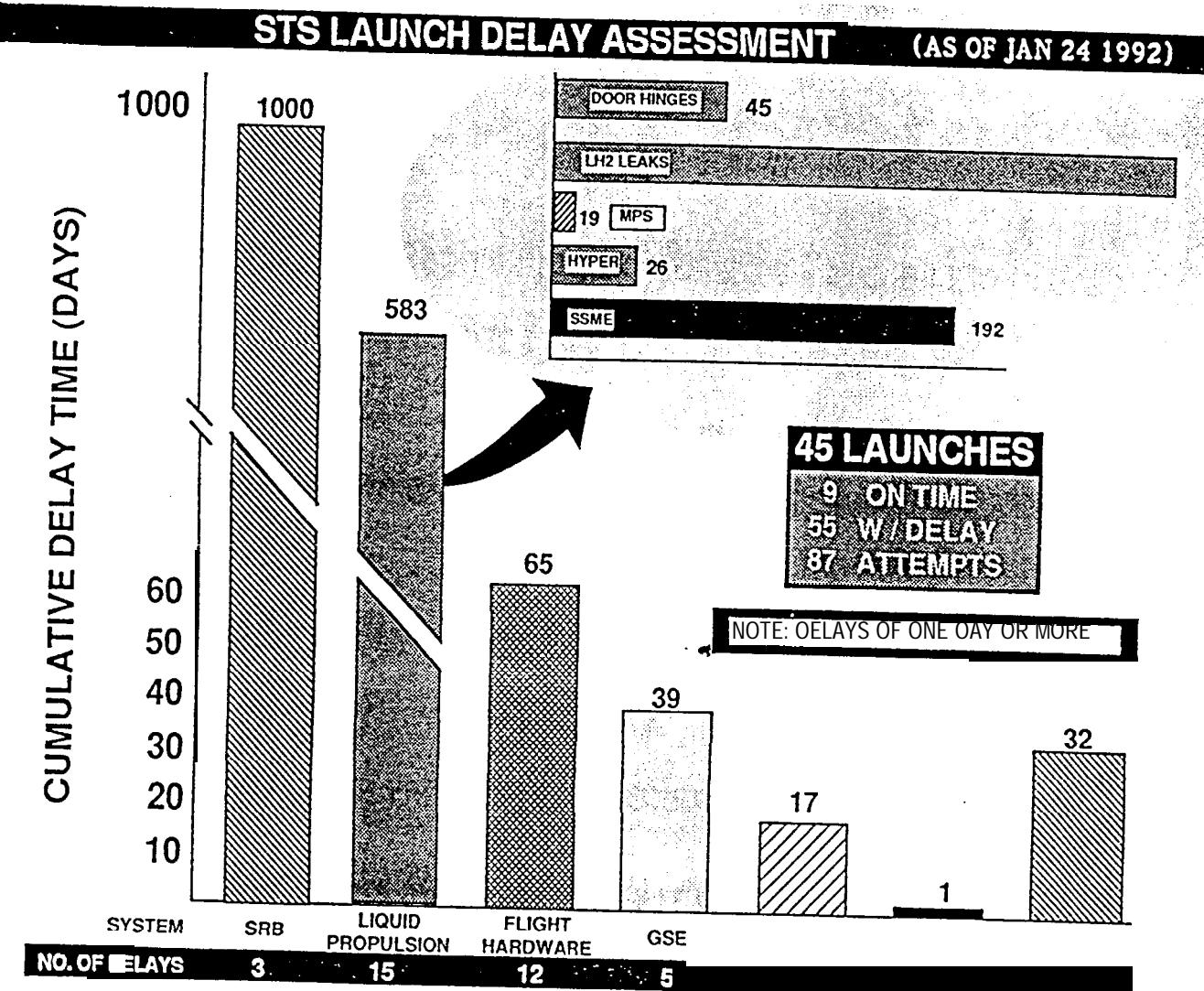
INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS

742 TOTAL FLIGHTS (1966-87), 58 failures



Where The Flight Failures Have Been In Launch Vehicles

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS



imstruments.

Laboratory did not exercise sufficient control over continuing changes in the spacecraft's design and its scientific

■ **Management flaw:** Project managers at the Jet Propulsion

space.

■ **Design flaw:** NASA engineers used technology that had been developed for operation in near-Earth orbit but was unsuitable for the more extreme conditions of interplanetary space.

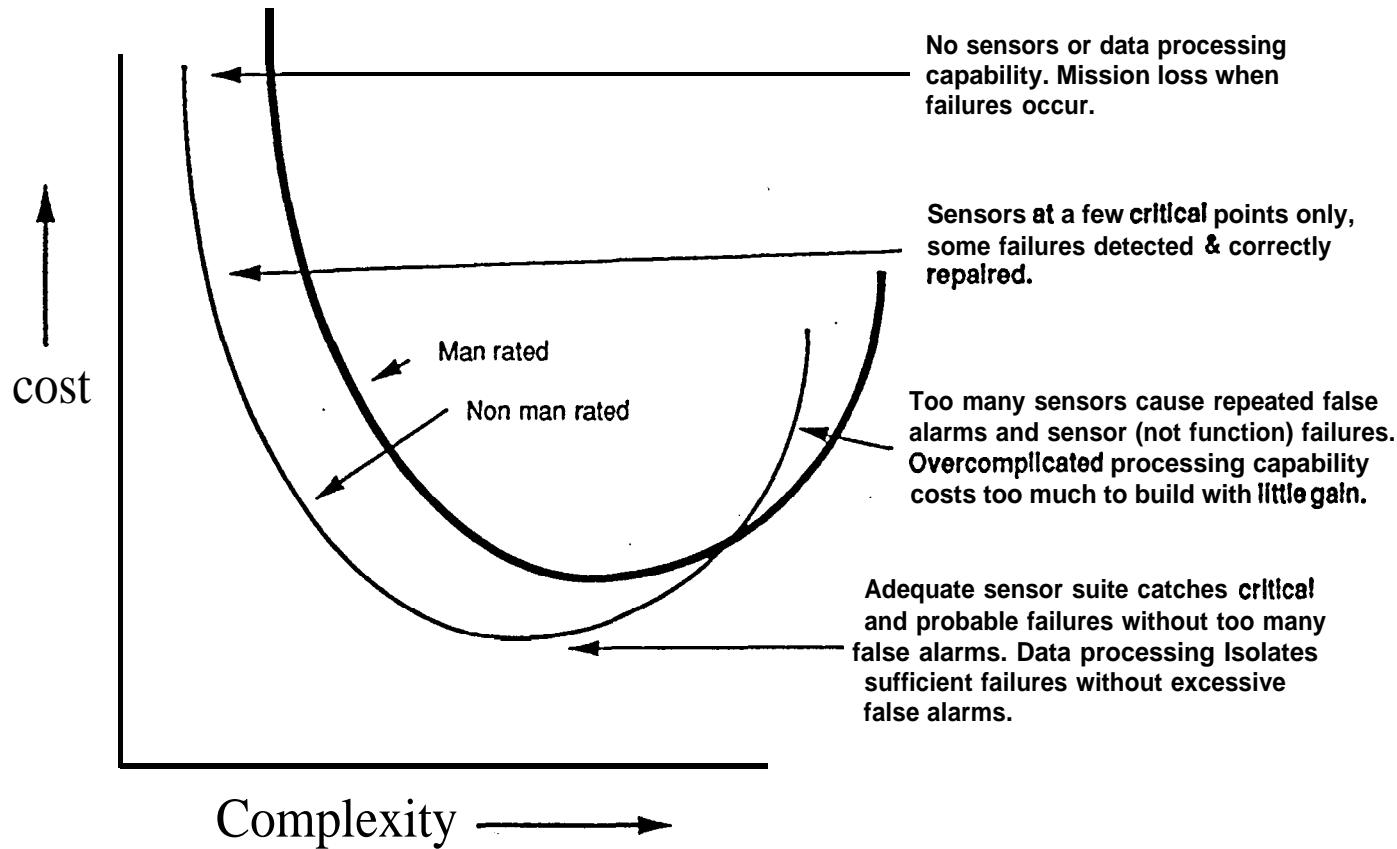
■ **Mechanical flaw:** A leak of volatile hydrazine fuel may have caused an explosion when the spacecraft's tanks were pressurized.



Troubled Spacecraft

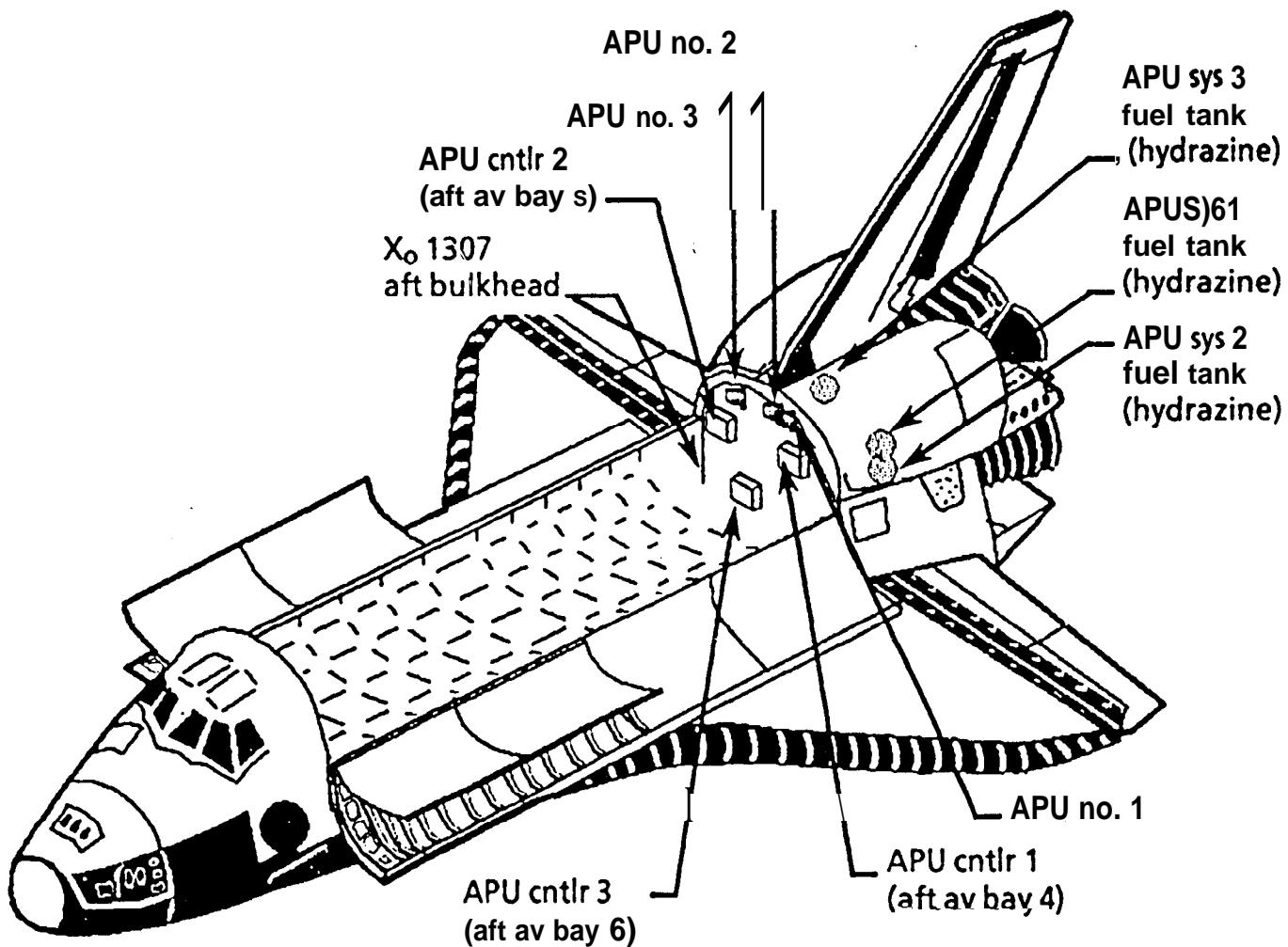
**Faillure of Mars Probe
Blamed on Fuel Leak**

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS



VHM COST OPTIMIZING CURVE

INTELLIGENT N'EUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS



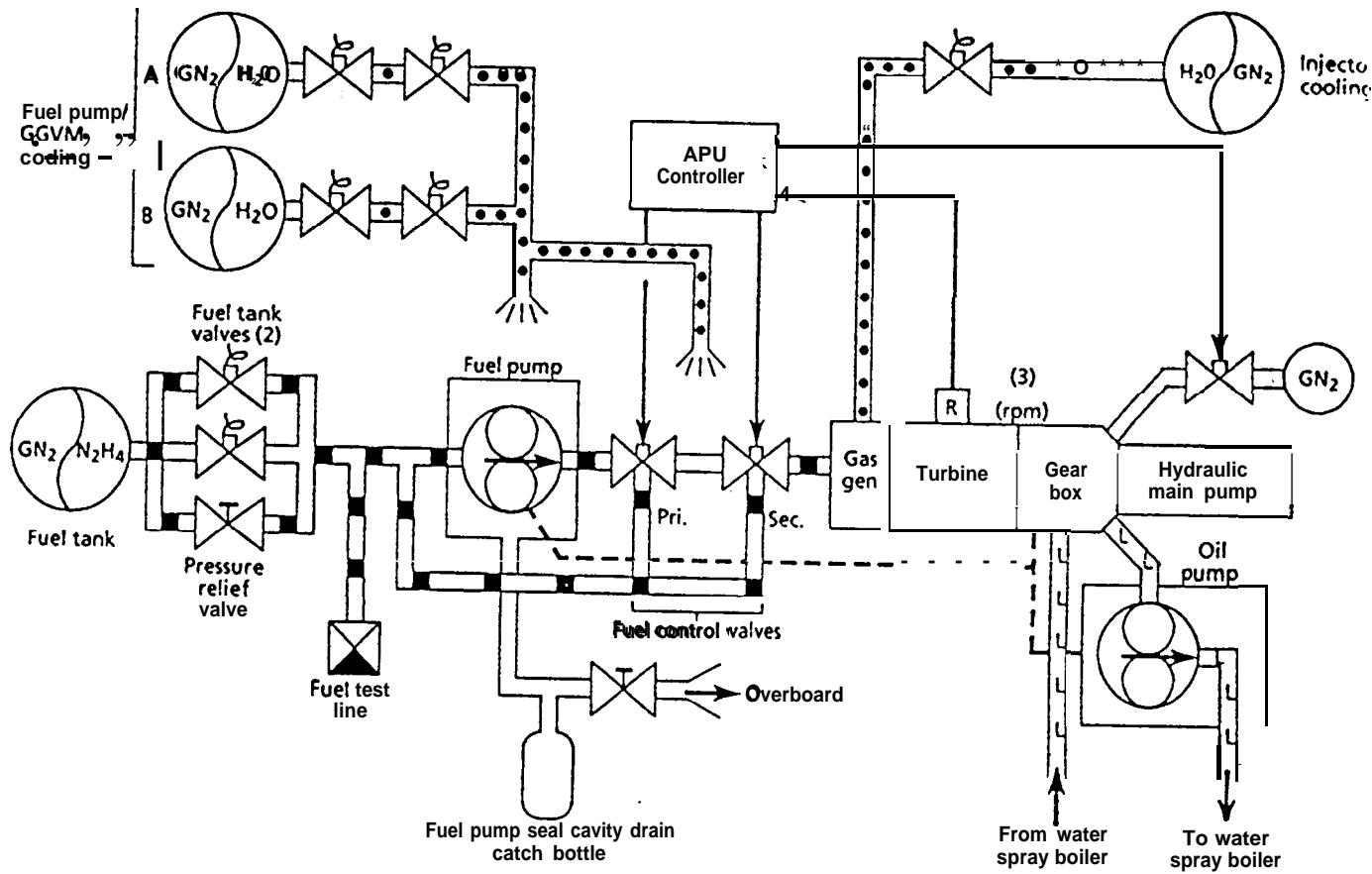
TARGET HMS - STS Auxillary Power Unit Location

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS

AUXILIARY POWER UNIT

- Provide power for the Orbiter hydraulic systems
 - liquid hydrazine -----> mechanical shaft power
- Hydraulic systems
 - actuate the Orbiter aerosurfaces
 - throttle and steer Orbiter main engines
 - deploy and steer landing gear
 - apply landing gear brakes
- Operation Cycle
 - t-5min to OMS-1 burn
 - deorbit burn and entry to just before landing

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS



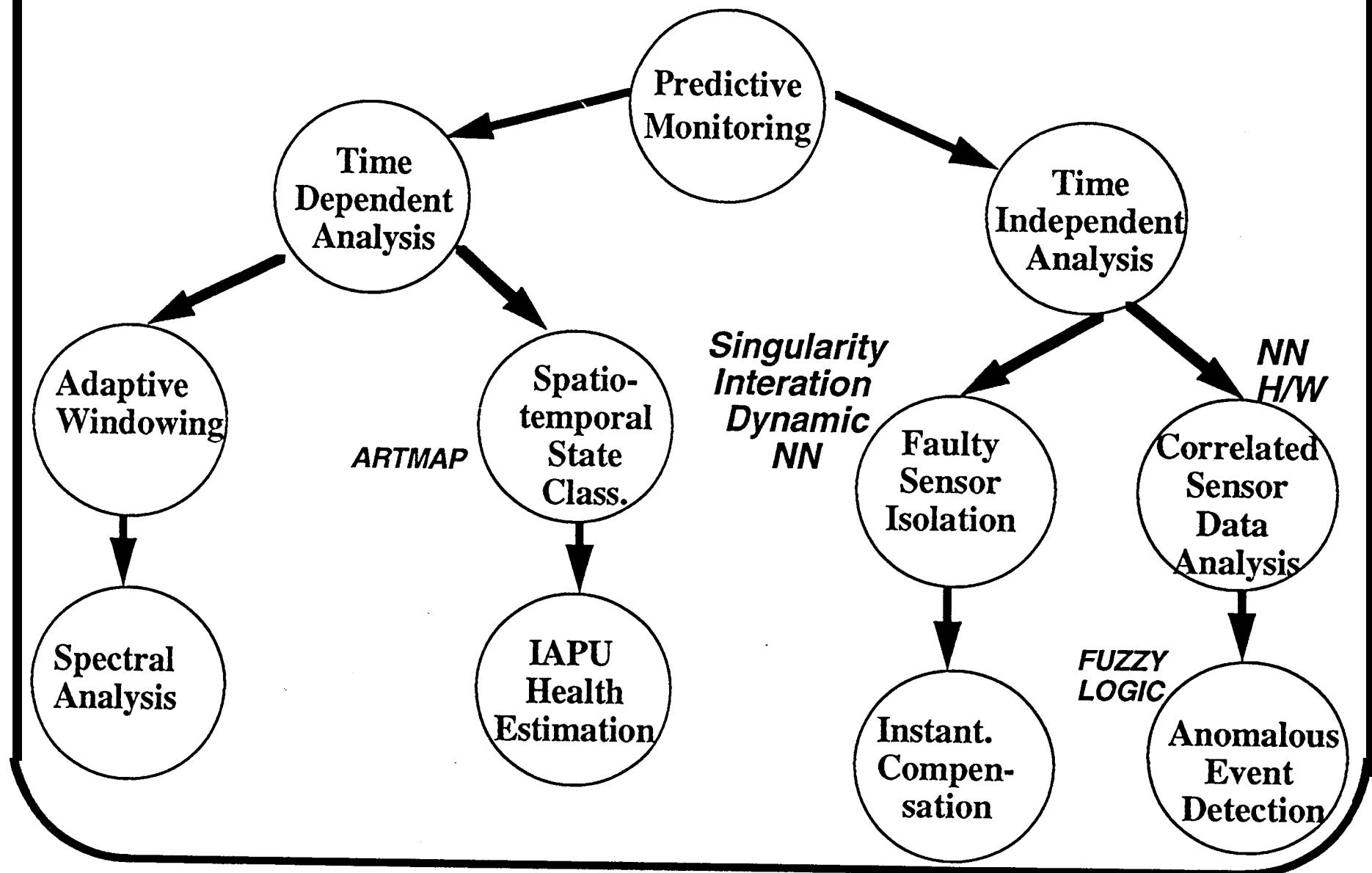
TARGET HMS - STS Auxillary Power Unit

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS

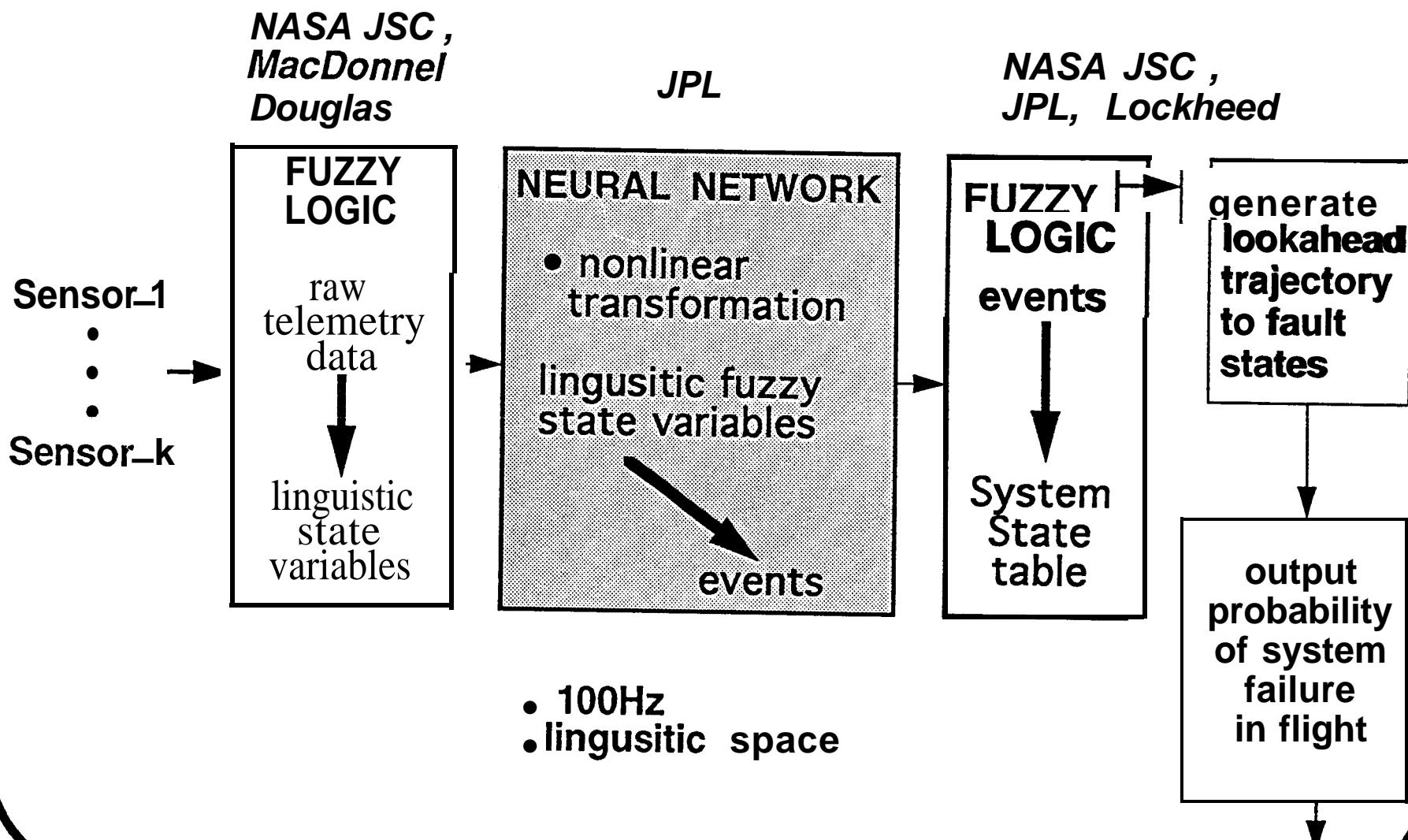
TECHNOLOGY ISSUES

- Engineering alarm limits - critical thresholds which define the acceptable range of engineering values on any telemetry channel
 - determined manually: hardcopy ISOE data, design information on spacecraft, rules of thumb
 - .Overreliance on domain experts leading to wide thresholds creating a range of undetected anomalies
 - monitoring of individual sensors via redlining approach
- Access only to snapshots of telemetry due to exploitation of low sensor acquisition rates. Further degradation due to noisy and incomplete data
- Specific diagnostics can be executed only if they were preconceived and preprogrammed
 - cannot currently correlate effects between multiple sensors in real-time
 - fault-detection to engine catastrophe time can be as short as 0.1 sec.

INTELLIGENT NEURO-FUZZY SYSTEM for STS APU Health Monitoring



Integration of Neural Networks & Fuzzy logic



INTELLIGENT NEURO-FUZZY SYSTEM for IAPU Health Monitoring

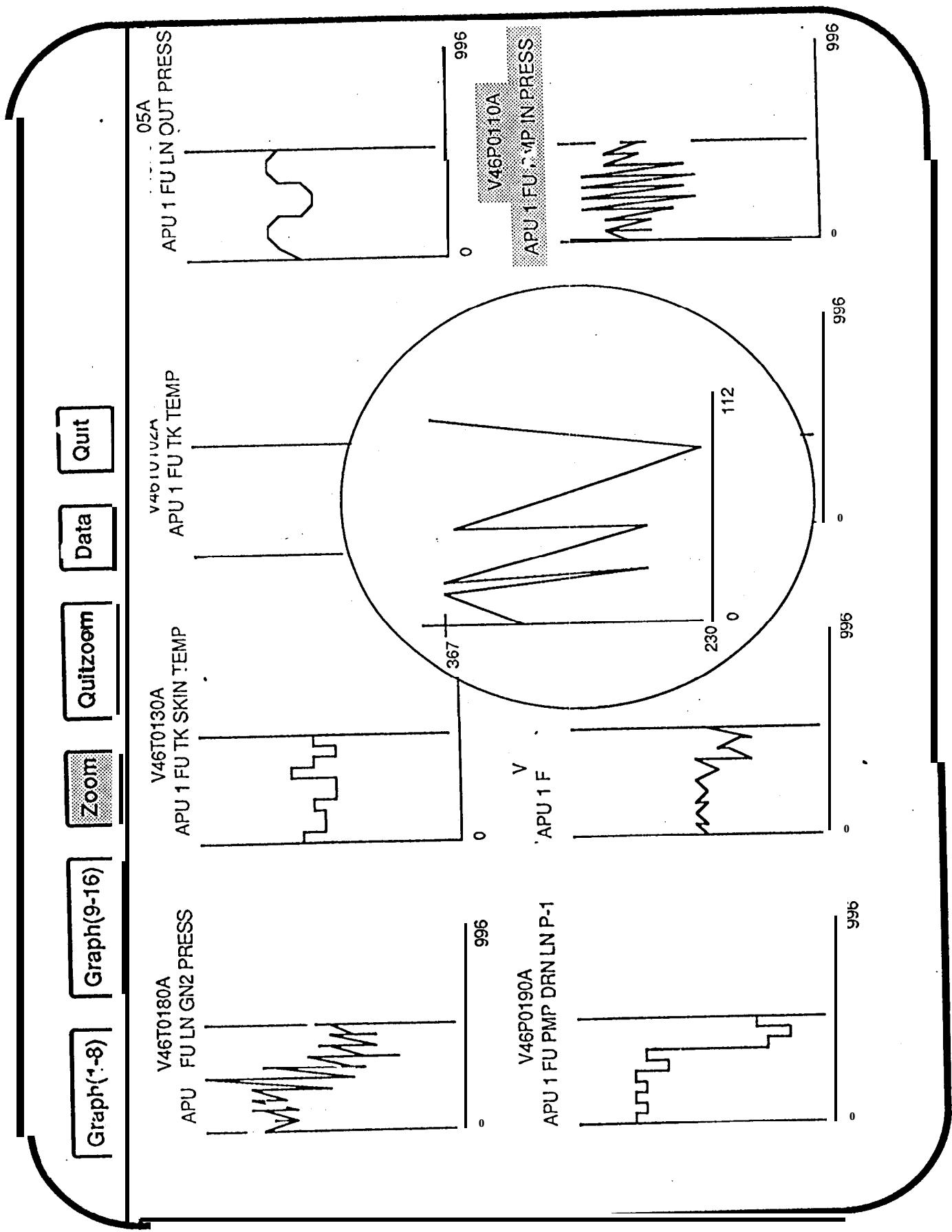
CAPABILITIES TARGETED FOR DEMONSTRATION: DEMO-1

- .detection of all red-line errors currently identified**
- .real-time correlation of data from multiple, heterogeneous sensor :**
 - faster-than-real-time anomaly propagation to determine probability of failure
 - both with (NN s/w) and without (NN h/w) time-lags
- .ease of augmenting expert-generated APU fault knowledge base without need to redesign the system**
- . isolating failed sensors as against failed subsystem / system**
 - reconstruct suspect information and minimize disruption of diagnostic process
- *** • synergistic integration of fuzzy logic and neural networks for real-time diagnostic applications**

INTELLIGENT NEUROPROCESSORS FOR LAUNCH VEHICLE HEALTH MANAGEMENT SYSTEMS

STS / IAPU HEALTH MONITORING

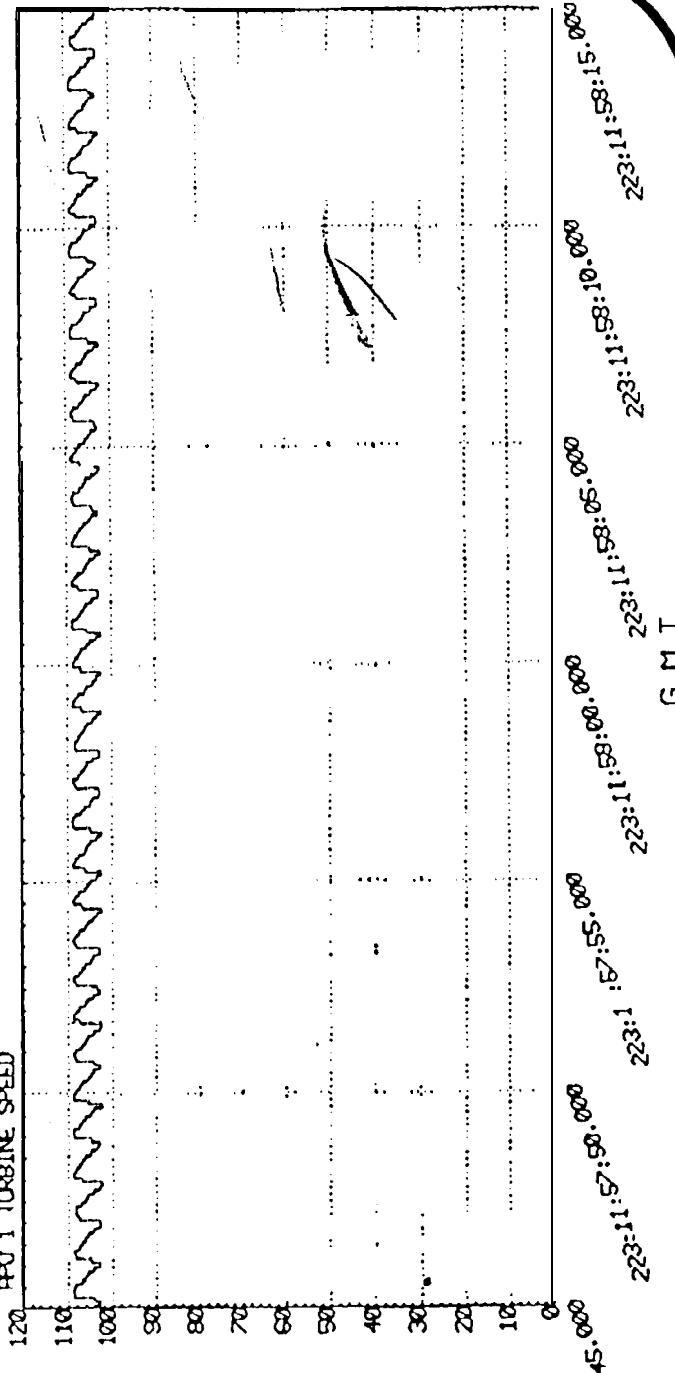
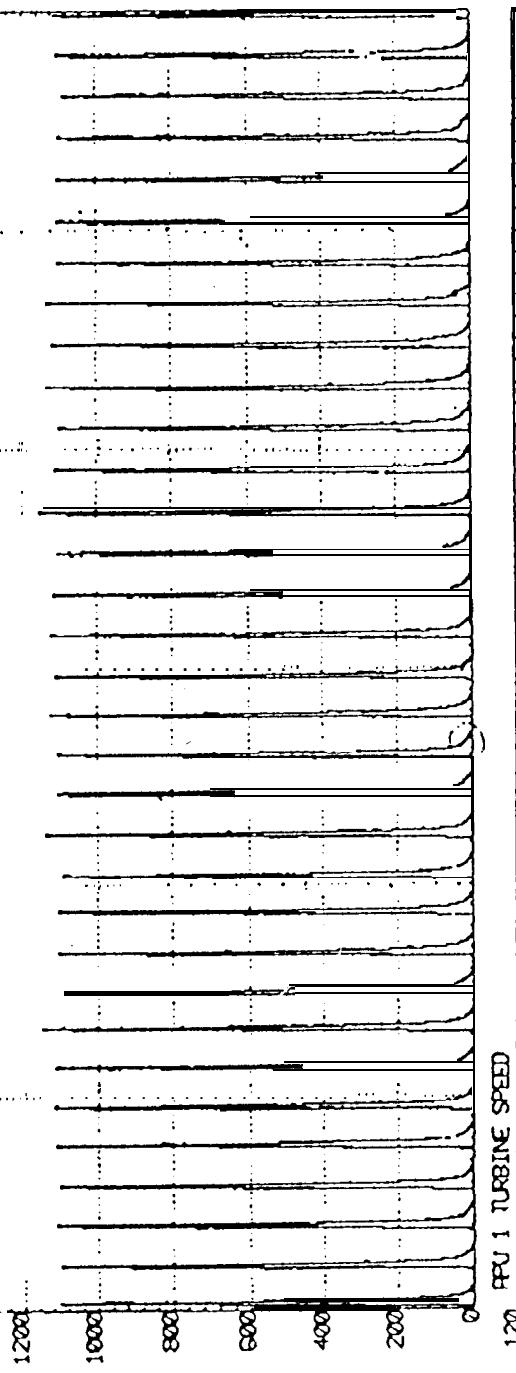
- Startup & mode-switch phases particularly difficult to monitor due to highly complex & nonlinear nature of IAPU dynamics
- reduced engine / **teststand** damage during test firings
 - = typically damage 1 APU every 2 weeks
- facilitate post-test diagnostic process
 - tool for APU knowledge engineering



SUBSYST: APU
STS-043

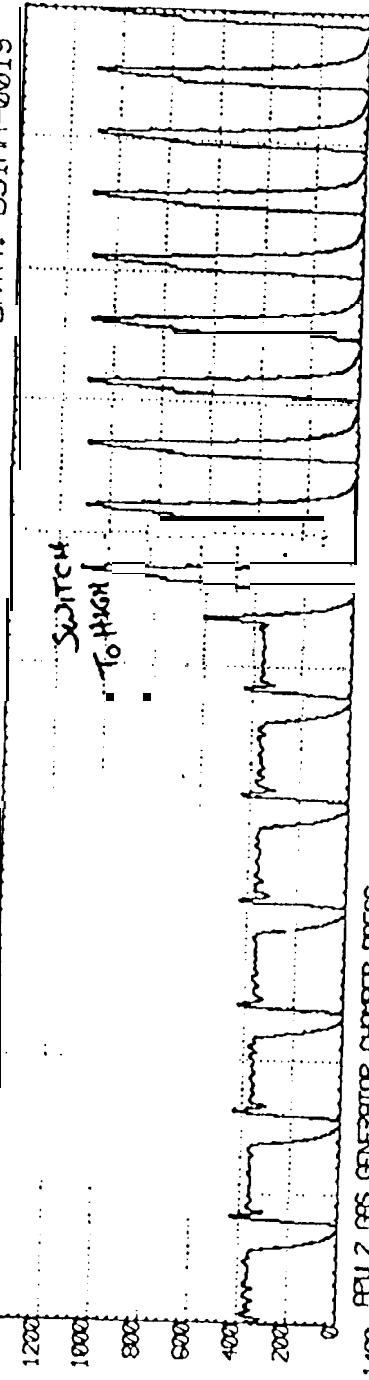
APU 1 CHAMBER PRESSURE VS TURBINE SPEED

FORMAT: APU1GG-SPD
DATA: DESGGPRES2

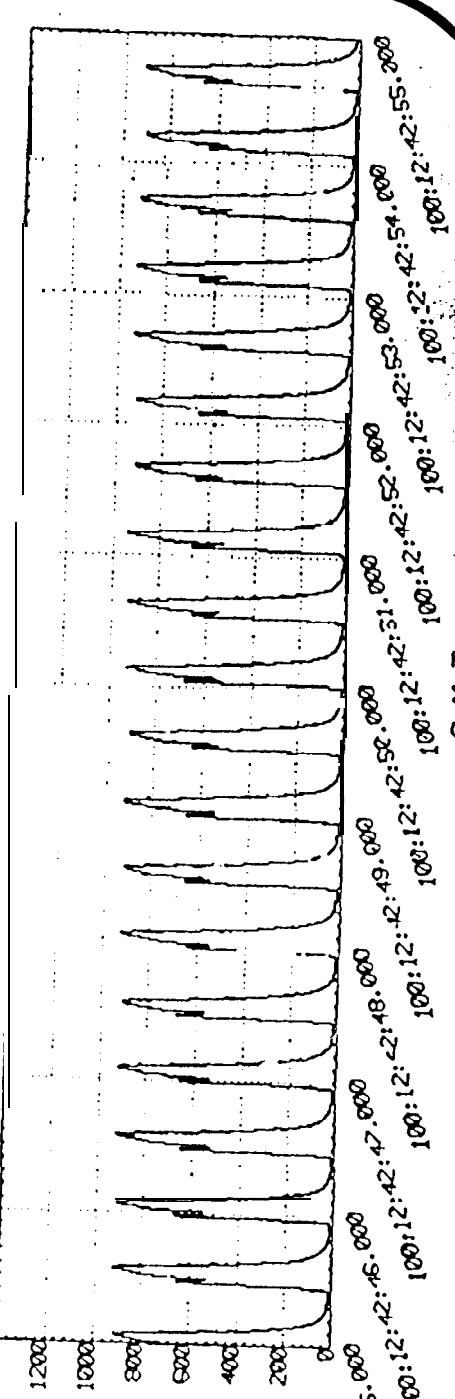
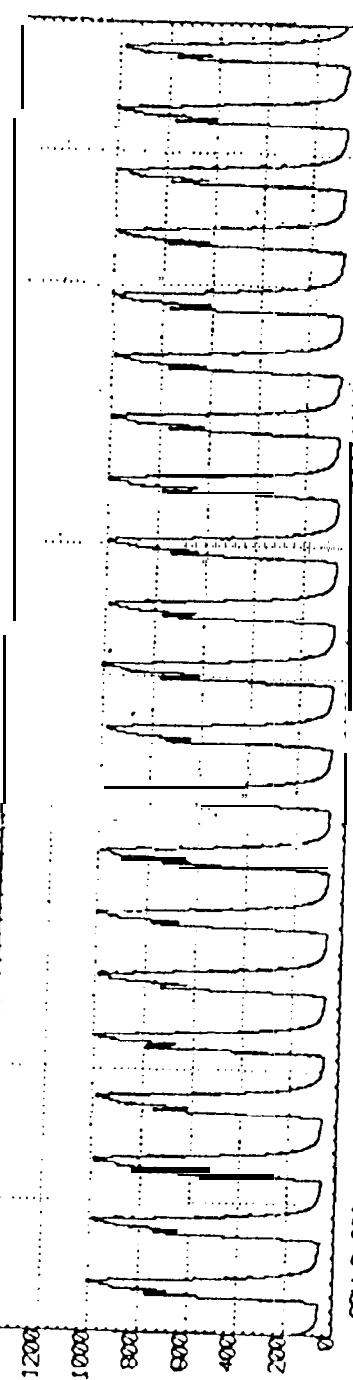


SUBSYSTEMS: PPU

APU CHAMBER PRESSURE
APU 1 GAS GENERATOR CHAMBER PRESS



FORMAT: APPUGPRESS
DATA: 831MR-0019

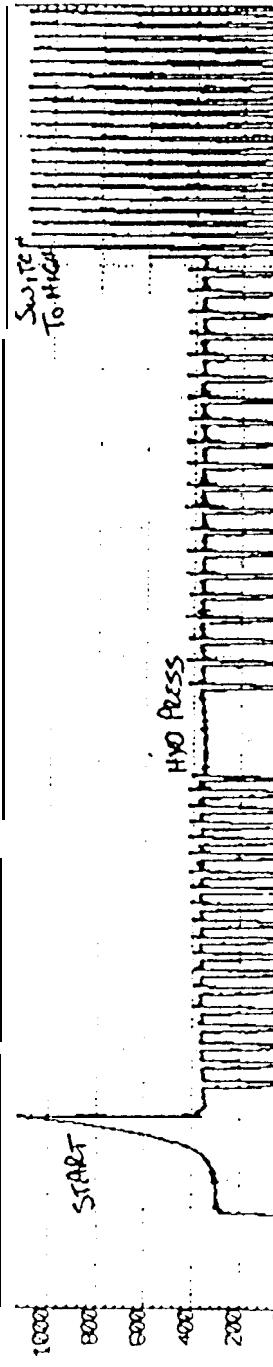


100:12:42:45.000 100:12:42:47.000 100:12:42:48.000
100:12:42:49.000 100:12:42:50.000 100:12:42:51.000
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100:12:42:55.000 100:12:42:56.000 100:12:42:57.000

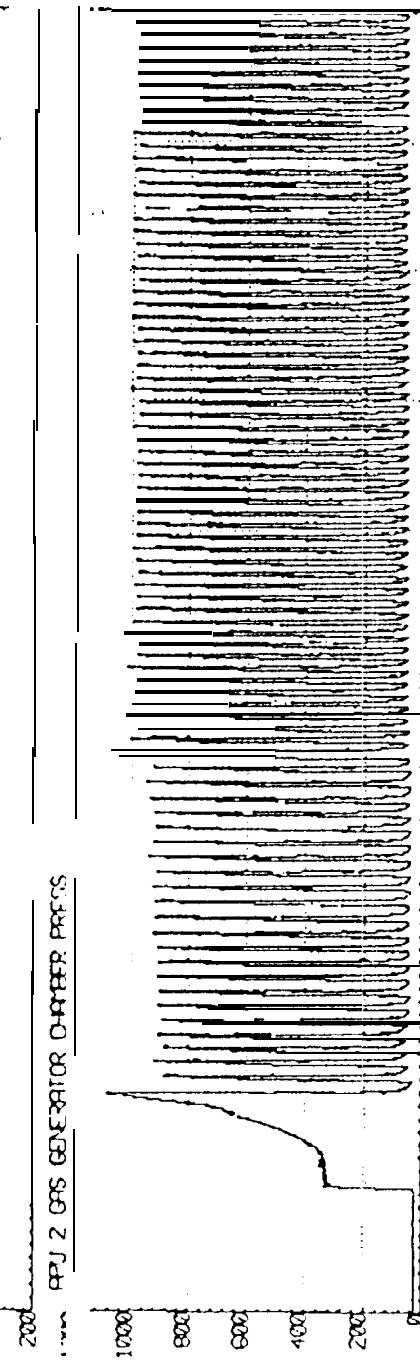
SUBSYSTEM: MER
STS-031

APPENDIX C: CHAMBER PRESURES

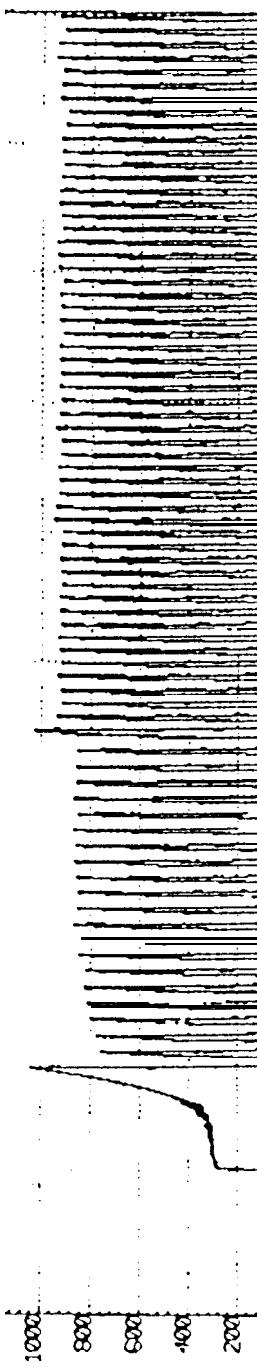
RUPA RAS GENEERIKAAT



FORMAT: EVT_PPU_3G
DATA: R21M9-0014



प्र० २ ग्रन्थ विवरण द्वारा प्रस.



THE USE OF GENETIC ALGORITHMS IN THE DESIGN OF COMPUTER PROGRAMS

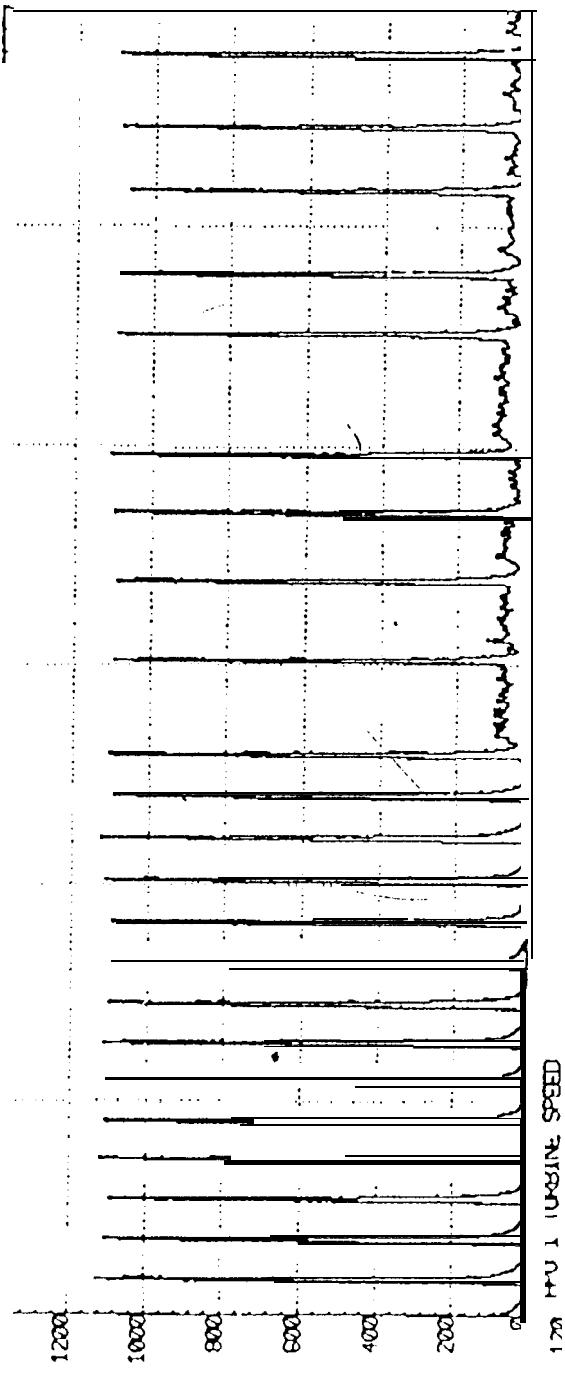
V4800000
(PS1A)

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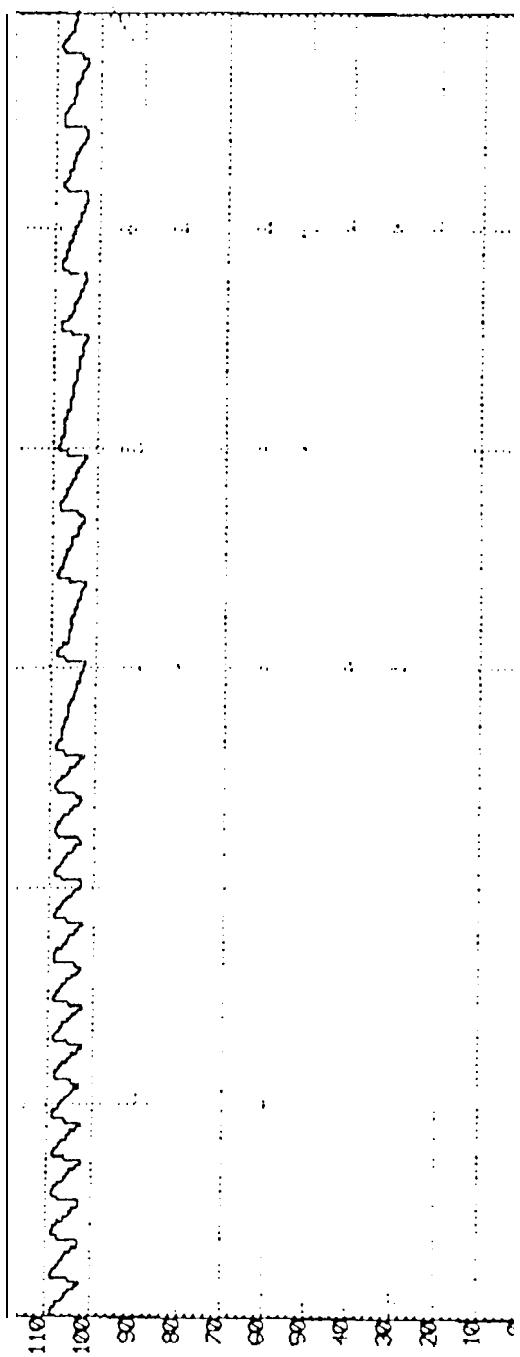
SUBSYSTEM: APU
STS-043

A PU 1 CHAMBER PRESSURE VS TURBINE SPEED
A PU 1 GAS GENERATOR CHAMBER PRESS

FORMAT: APU1GG-SPO
DATA: DESGPREFS?



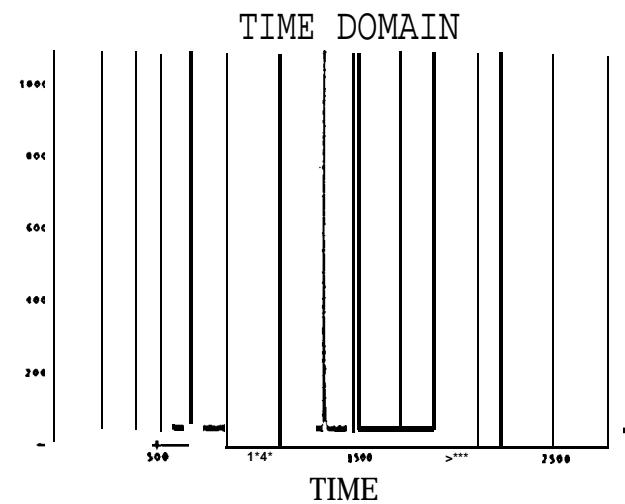
V46P01209
(PSIA)



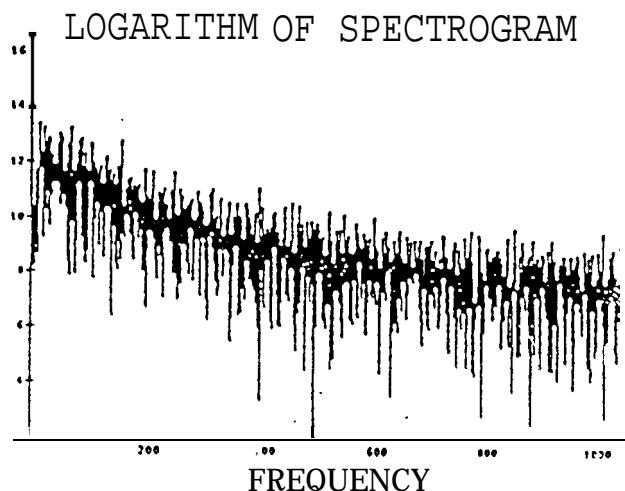
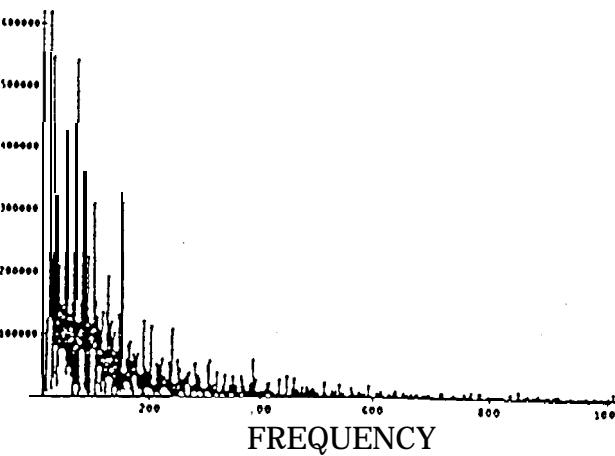
V46P011359
(PSI)

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223:11:58:45.000
G M T

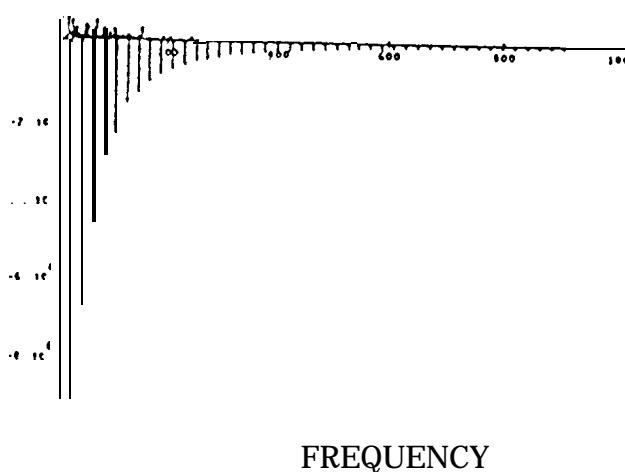
VHM SENSOR DATA WITH CHANGING FREQUENCY AND ADDITIONAL GROUND NOISE



SPECTROGRAM

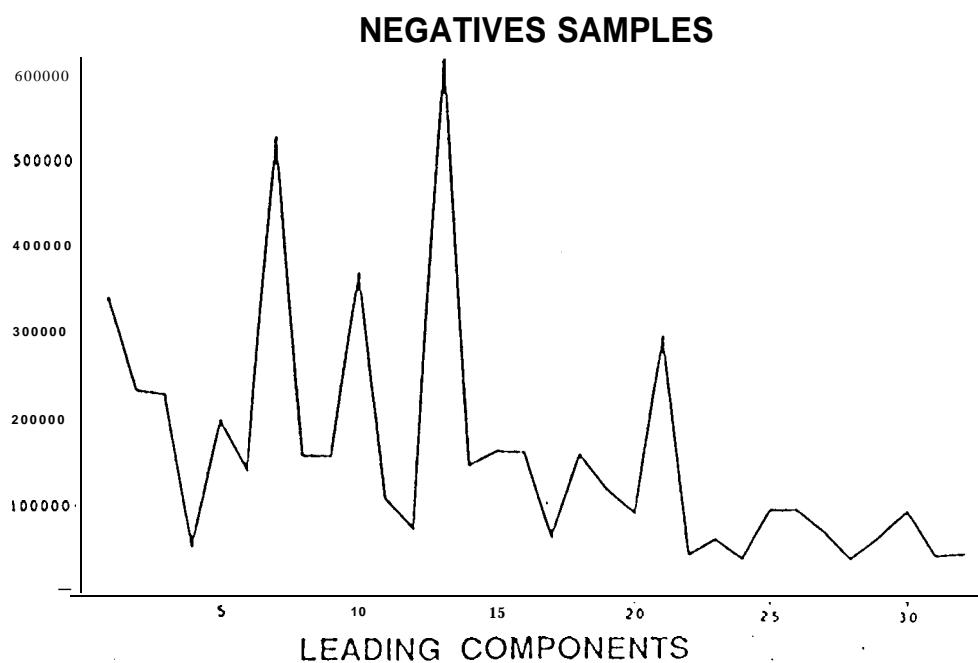
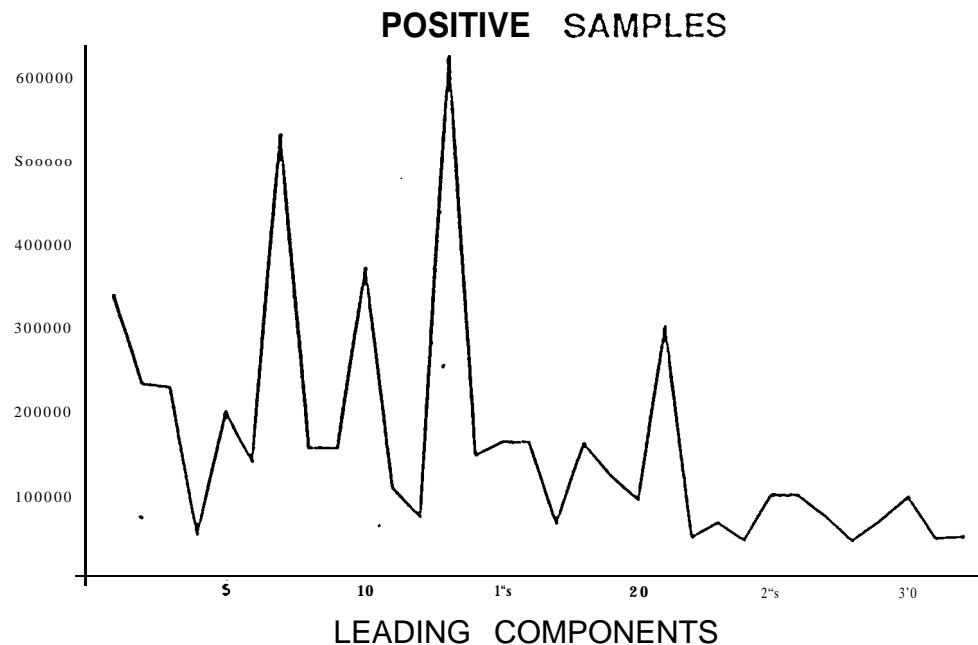


SPECTROGRAM DIFFERENCE



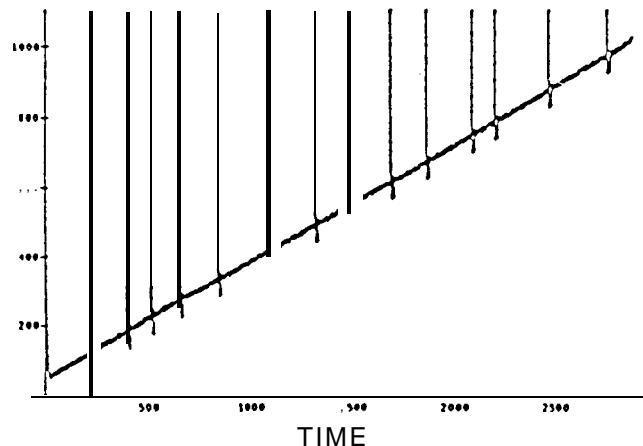
SAMPLED SPECTROGRAM DIFFERENCE

VHM SENSOR DATA WITH VARIATIONS IN FREQUENCY AND GROUND NOISE

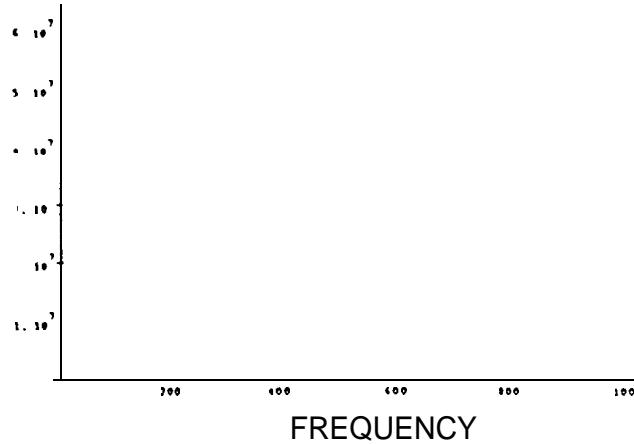


VHM SENSOR DATA WITH CHANGING FREQUENCY\\ AND NOISE BUILDUP

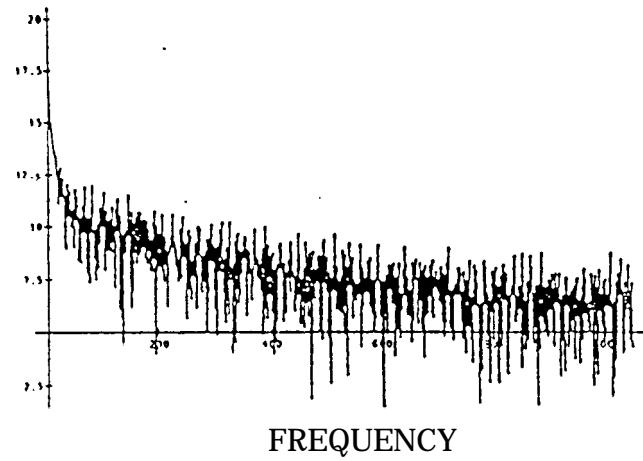
TIME DOMAIN



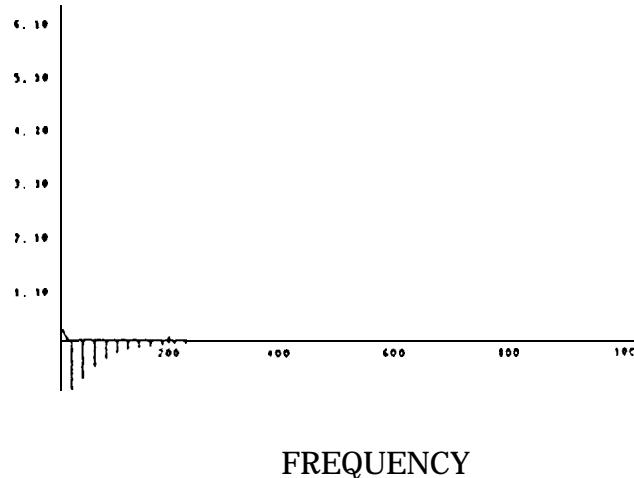
SPECTROGRAM



LOGARITHM OF SPECTROGRAM



SPECTROGRAM DIFFERENCE



SAMPLED SPECTROGRAM DIFFERENT

VHM SENSOR DATA WITH VARIATIONS
IN FREQUENCY AND BUILDUP NOISE

